

Surface Characterization

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Abstract—The combined PROSPECT leaf optical properties model and SAIL canopy bidirectional reflectance model, i.e. PROSAIL, has been used for about fifteen years to increase our understanding of plant canopy spectral and bidirectional reflectance in the solar domain and to develop new methods of vegetation biophysical properties retrieval. It links the spectral variation of canopy reflectance with its directional variation. This link is the key to simultaneously estimate biophysical/structural canopy variables for applications in agriculture, plant physiology, and forestry at different scales. PROSPECT and SAIL are still evolving: they have undergone recent improvements both at the leaf and the plant levels and became one of the most popular radiative transfer tools in these domains due to their ease of use, their robustness, and because they have been validated by many lab/field/space experiments over the years. This paper is intended to review this subject, which has been extensively researched in optical remote sensing.

Keywords—canopy spectral and directional reflectance; models; PROSPECT; SAIL; PROSAIL;

I. INTRODUCTION

From the beginning of optical remote sensing, radiative transfer models served to help understand how light is intercepted by plant canopies and to interpret the reflectance in terms of vegetation biophysical characteristics. Since they describe more or less precisely the two main physical principles involved in such a medium, i.e. absorption and scattering, one is apt to believe that they can be very useful in tasks such as designing vegetation indices, performing sensitivity analyses, and building up inversion procedures destined to accurately retrieve these characteristics. Among all the codes published in the literature during the last two decades

[1], the SAIL canopy bidirectional reflectance model and the PROSPECT leaf optical properties model are the most popular codes and are looked on as standards. Their combination in PROSAIL about fifteen years ago linked the spectral variation of canopy reflectance, which is mainly related to leaf biochemical contents such as total chlorophyll, water, or dry matter, with its directional variation, which is classically related to the canopy architecture, i.e. leaf area index, leaf angle distribution, hot spot parameter or clumping effect. This link is essential to simultaneously estimate these variables. It partly explains the success of PROSAIL.

II. MODEL OVERVIEW

Now in widespread use in the remote sensing community, SAIL (Scattering Arbitrary Inclined Leaves) was one of the earliest canopy reflectance models [2-3]. It was an extension to the 1-D model developed by [4] and solved the scattering and extinction of four upward/downward fluxes within the canopy to predict the bidirectional reflectance of homogeneous plant canopies. SAIL has given rise to several versions: SAILH which includes the hot spot parameter after a formalism by [5], FLSAIL [6] and FluorSAIL [7] which take into account the chlorophyll fluorescence effects on reflectance, GeoSAIL which is the combination of SAIL and a geometric model to simulate discontinuous canopies [8], and recently 4SAIL which is a numerically robust and speed-optimized version of the model [9]. SAIL was successfully compared to other 1-D and 3-D codes in the frame of the RAMI (Radiation transfer Model Intercomparison) organized by the JRC in Ispra, Italy.

At the leaf level, PROSPECT pioneered the simulation of hemispherical reflectance and transmittance of various green monocots and dicots, and senescent leaves, over the solar

spectrum [10]. The latter are represented as one or several absorbing plates with rough surfaces giving rise to isotropic diffusion. The model uses two classes of input variables: the leaf structure parameter N , number of compact layers specifying the average number of air/cell walls interfaces within the mesophyll, and the leaf biochemical content which has varied over the years [11-13]. The directional optical properties have been recently introduced by [14] and a new version including chlorophyll a fluorescence is underway [15].

Where other models have been developed, generally they have received inadequate validation and comparison with existing ones, either due to lack of resources or limited access. The large diffusion of PROSPECT and SAIL is attributed to their simplicity, accuracy, and above all their availability to the research community.

III. COUPLING OF PROSPECT AND SAIL

The first model inversions of SAIL performed by [16] allowed estimates of canopy architecture (LAI, leaf angle distribution) by using bidirectional reflectance measurements acquired in a given waveband. Currently, only POLDER and MISR have the capability to produce such data. Most sensors measure the Earth radiance in several wavebands and one direction. However, multispectral or hyperspectral data cannot be inverted by SAIL alone because the increase in the number of wavebands rapidly leads to an under-determined system. Since leaf reflectance, leaf transmittance, and soil reflectance are three wavelength dependent input variables of SAIL, the implementation of this model to predict biophysical variables from canopy reflectance spectra at given solar and viewing angles in a defined relative azimuth plane requires at least three times as many variables as wavelengths. As a consequence, the inversion of SAIL is impracticable unless several viewing angles are available. In order to reduce the dimensionality of the inverse problem and to assess the canopy biochemistry, SAIL was coupled with PROSPECT early in the 90's to give PROSAIL. The main input variables of the latter are shown in Table I. This was the beginning of a long series of published literature.

TABLE I. MAIN VARIABLES OF PROSAIL

Symbol	Quantity	Unit
N	Leaf structure parameter	-
C_{ab}	Chlorophyll a+b concentration	$\mu\text{g cm}^{-2}$
C_w	Equivalent water thickness	cm
C_m	Dry matter content	g cm^{-2}
C_b	Brown pigments content	-
LAI	Leaf Area Index	-
LIDF	Leaf Inclination Distribution Function	-
S_1	Hot spot parameter	-
$\rho_s(\lambda)$	Soil reflectance assumed Lambertian	-
SKYL(λ)	Ratio of diffuse to total incident radiation	-
sza	Solar zenith angle	deg
vza	Viewing zenith angle	deg
raa	Relative azimuth angle	deg

It should be mentioned that PROSPECT has also been integrated into other canopy reflectance models: FCR [17], NADI [13, 18], DART [19], GeoSAIL [20-23], FLIM [24], SPRINT [25], and FLIGHT [23].

IV. SIMULATIONS WITH PROSAIL

Models can be run in direct mode. One of the first applications of PROSAIL was the study of the red edge [26]. Spectral shifts in this wavelength domain were shown to be mainly produced by variations in C_{ab} and LAI and the position of the inflexion point to be almost insensitive to soil substrate and atmospheric conditions. By successively changing the input variables, [27] performed a rough sensitivity analysis which revealed that N slightly changed the canopy reflectance and that LAI and the average leaf angle of the LIDF produced quite similar effects in the model. As a result, the separate quantification of these two variables is problematic. The canopy reflectance sensitivity to leaf optical properties was evaluated by [28] who showed that the leaf biochemical signal could be enhanced at the canopy level by a factor of two. Finally, [29] and [30] compared modeled and measured spectra to monitor changes in sugar beet crop spectral reflectance caused by diurnal water stress.

Several vegetation indexes have been also screened or designed using PROSAIL: for instance, the relationship between the Weighted Difference Vegetation Index and the LAI was verified for varying external factors [31]. Eight vegetation indexes were compared with a neural network approach to determine the canopy gap fraction [32]. References [24, 33] simulated canopy reflectance spectra in the VIS-NIR to test the Transformed Chlorophyll Absorption in Reflectance Index / Optimized Soil-Adjusted Vegetation Index ratio which was supposed to be sensitive to low chlorophyll values and resistant to vegetation non-photosynthetic materials. In the SWIR, the fuel moisture content at canopy level was related to the Simple Ratio Water Index [34] and to the Normalized Difference Water Index [20]. These two studies pointed out the difficulty of accurate estimations using a semi-empirical approach and demonstrated the need for a coupled leaf-canopy model to successfully estimate C_w at canopy level.

Model simulations can also help quantify the contribution of canopy input variables one-by-one to canopy reflectance. Recent studies based on statistical methods like the Design Of Experiments for Simulation [35-36] or the Extended Fourier Amplitude Sensitivity Test [20] made it possible to perform complete sensitivity analyses of PROSAIL. By quantifying the contribution of each of the input variables to the model outputs, as well as their interactions, such analyses proved informative about the optimal wavebands and viewing directions to retrieve canopy biophysical characteristics. For instance, the canopy reflectance turned out to be the most sensitive to C_{ab} in the backward direction where shadows are reduced, to LAI in the forward direction where photons have interacted twice with the plants and to leaf orientation at nadir where the soil background is visible [35-36]. The spectral domain where canopy reflectance is sensitive to water or to the specific leaf area was also found this way [20]. A further integrated model including the soil BRDF and the atmosphere showed how top-of-atmosphere hyperspectral radiances under multiple view angles could be predicted [37]. This model was used to assess the performance of several mission concepts by sensitivity analysis. Finally, one can mention the availability of FluorMODgui, a Graphic User Interface software package [38] developed within the frame of the FluorMOD project funded

by the European Space Agency and which enables simulation of leaf and canopy reflectance with or without the effects of chlorophyll fluorescence [7].

V. INVERSION OF PROSAIL

The inverse problem is crucial in remote sensing where the radiometric signal has to be interpreted in terms of canopy biophysical characteristics. Although they represent a majority of approaches to retrieve these characteristics, we will omit here any reference to vegetation indices. One can attack the inverse problem of PROSAIL in three ways: iterative methods, look-up tables, and neural networks. The first two methods share the minimization of the distance between simulated and measured reflectance, while the last one focuses on the biophysical variables.

The first attempt to iteratively estimate C_{ab} , C_w and LAI from high resolution reflectance spectra acquired at nadir in the solar domain was made on sugar beet crops [39]. While retrieving the main canopy variables using AVIRIS, CASI, and TM sensors, unstable inversions were observed due to interferences between the LAI and the LIDF in the reflectance signal, but the determination of total chlorophyll ($C_{ab} \times LAI$) or water ($C_w \times LAI$) contents seemed to be attainable [13, 27, 40, 41]. A limited number of wavebands and view angles should however permit a better separation of the variables [42-44]. The main problem of iterative inversion methods is a prohibitive computation time and the risk of converging on a local minimum.

The look-up table is conceptually a simple technique to overcome the above-mentioned limitations. It consists of generating an output table with PROSAIL for a discrete set of input variables covering their natural range of variation. This table is then used to find the solution to the problem, i.e., the sets of input variables which provide the closest estimate of reflectance. This method permitted estimation of C_{ab} , LAI, fAPAR, and fCover [21, 45-47].

As an alternative, neural networks have lately become a standard method to operationally invert PROSAIL. The input variables are drawn at random within given distribution laws to produce a learning data set to train the neural network and a test data set to evaluate its performance. This approach has been successfully applied to VEGETATION [48], MERIS [49] and POLDER [50] to determine the LAI or the gap fraction.

The three methods were adapted to account for prior information and evaluated over synthetic canopy reflectances [51]. Results demonstrated a significant improvement of canopy variables retrieval when using such information.

VI. CONCLUSION

As illustrated in this paper, plant canopy reflectance models have become essential tools for the analysis of optical remote sensing data, providing meaningful links between radiometry for both ecology and precision agriculture. The coupling of PROSPECT with SAIL made it possible to physically interpret multispectral or hyperspectral images in terms of leaf biochemical content, but also, when the requirements are fitted, in terms of canopy architecture. When run in direct mode,

PROSAIL provides a means to generate databases and test new spectral indexes, to perform sensitivity analyses which will allow to better design forthcoming sensors devoted to specific applications. When embedded in an inversion procedure, it turns into a powerful tool to derive new products. The first maps of chlorophyll or water content appeared just a few years ago. In the future, terrestrial ecosystems could be monitored in a totally different way when we are able to explore the main photosynthetic pigments separately or leaf structural properties.

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